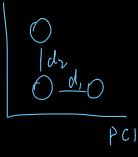
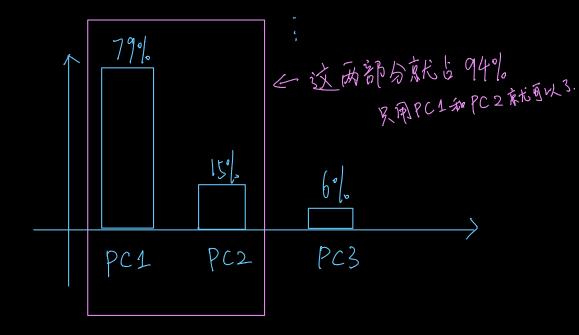
## PCA:

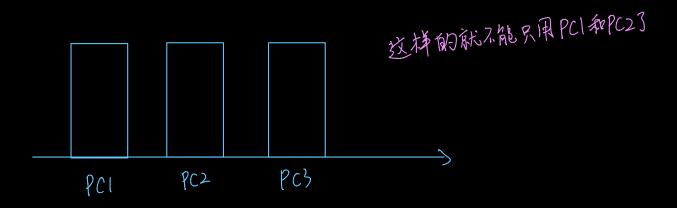
· PCI 此PC2重要,所以 0 1d2 0 d10



3 di = dr At, di value more than dr

· Variation for PC1 = SS (distances for PC1)





## PCA AtaB:

pca = procomp (traindata[, c(3:9)],

center = TRUE,

Scale = TRUE)

goal: draw a graph that shows how the samples are related (or not related) to each other.

proompl) return 3 thing:

· × - PCs for drawing a graph.

plot (pca \$x[,1], pca \$x[,2])
横轴 纵轴

· S dev — standard deviation, to calculate how much Variation in the original data each PC accounts for.

library (99 plot 2)

PCa. data  $\leftarrow$  data. frame (Sample = rownames (pca \$x),  $\leftarrow$  1 col with sample ids X = pca \$X [, 1], \$1 < 2 cols for X, Y coordinates X = pca \$X [, 2], \$1 < 2 cols for X, Y coordinates

- for fire of

data frame

ggplot (data = pca.data, aes (x=X, y=Y, label = Sample)) +

geom-text() + < plot labels (text) rather than "dot" foreg.

xlab (paste ("PC1-", pca.var.per[1], "%", sep="")) +

ylab (paste ("PC2-", pca.var.per[2], "%", sep="")) +

theme\_bw() + < graph bg white

gg title ("My PCA Graph") < the title of the graph

## · rotation

loading\_scores < pca \$ yotation [, 1].

The prcomp() function calls the loading scores "rotation"

PcaScore 

abs[loading\_scores]

最左边的有很大的负值.

最右边的有很大的正值。

abs() - sort based on the number's magnitude, rather than from high to low.

pca Score Rank < sort (pca Score, decreasing = TRUE)

pca\$votation[pcaScoreRank,1] 显示scores(第十一名)